

1 Supplementary Materials for the paper entitled:

2 Enhanced PM_{2.5} pollution in China due to aerosol-cloud
3 interactions

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16 **1. Model evaluation**

17 The model predictions agree fairly well with surface meteorological observations. The MBs in
18 January/July are 0.79/0.42 m s⁻¹, -1.28/-1.17 K, -0.87/-1.06 g kg⁻¹, 2.3/11.6 mm month⁻¹ for
19 WS10, T2, Q2, and precipitation, respectively. Emery et al.¹ proposed benchmark values for
20 satisfactory performance for WS10, T2, and Q2: MB within ± 0.5 m s⁻¹, GE ≤ 2.0 m s⁻¹, RMSE
21 ≤ 2.0 m s⁻¹ and IOA ≥ 0.6 for WS10, MB within ± 0.5 °C, GE ≤ 2.0 °C, and IOA ≥ 0.8 for T2,
22 and MB within ± 1.0 g kg⁻¹, GE of ≤ 2.0 g kg⁻¹, and IOA ≥ 0.6 for Q2. We note that these
23 benchmark values are proposed based on the performance of a series of model simulations with
24 four dimensional data assimilation (FDDA). Nevertheless, FDDA is not utilized here to allow
25 full aerosol-cloud-radiation interactions. Therefore, the model performance is not expected to be
26 as good as those with FDDA. Table 2 in the main text shows that the performance statistics for
27 WS10 in July and Q2 in January fall within benchmark ranges, and those for Q2 in July are very
28 close to the benchmark ranges. The WS10 in January and T2 in both months exceed the
29 benchmark range but still have smaller or similar biases compared with most previous WRF-
30 Chem applications without FDDA over East Asia²⁻⁷, in which MBs range from 0.4 to 3.1 m s⁻¹
31 (mostly 1.2 to 2.6 m s⁻¹) for WS10 and from -1.8 to 1.0 K (mostly -1.8 to -0.8 K) for T2,
32 respectively. Therefore, the model performance is considered to be decent.

33 With regard to surface air quality, the model performance for PM_{2.5} has been described in the
34 main text. For gaseous pollutants, the model-measurement agreement is decent for NO₂ and daily
35 maximum O₃ concentrations in both months, and SO₂ concentrations in January, with NMBs
36 within $\pm 20\%$. The model overestimates SO₂ concentrations in July, likely attributable to the
37 uncertainty in emission inventory and insufficient treatment of SO₂ oxidation reactions on dust
38 surface⁸ and on fine aerosols with high relative humidity and NH₃ neutralization^{9,10}. The

39 observational data of PM_{2.5} chemical components are quite sparse and not publicly available
40 during the simulation periods. In this study, we compare with chemical component observations
41 obtained during a field campaign period (from July 22nd-31st, 2013) at two sites located in the
42 North China Plain (see Supplementary Figure 1). The comparison results are shown in
43 Supplementary Figure 3. Simulated PM_{2.5} concentrations agree fairly well with observations;
44 NMBs are within ±6% for both sites. As for chemical components, NO₃⁻ concentration is
45 overestimated (NMB = 6% to 52%), while SO₄²⁻ concentration is underestimated (NMB = -37%
46 to -63%). There is a good agreement for NH₄⁺ (NMB within ±23%) and total SNA (sulphate-
47 nitrate-ammonium, NMB within ±23%). The overestimation of NO₃⁻ and underestimation for
48 SO₄²⁻ are consistent with previous studies over East Asia, probably attributed to the lack of some
49 chemical formation pathways in the modeling system ^{8,10,11}. Simulated elemental carbon (EC)
50 concentrations approximately double observed EC concentrations. EC concentrations are
51 strongly affected by local emissions, while the spatial distribution of our emission inventory
52 (36 km × 36 km) may not be able to capture local emission sources surrounding observational
53 sites, leading to model-observation bias. The overestimation may also be attributable to the
54 absence of EC aging in WRF-Chem, which leads to reduced fraction of hydrophilic EC and thus
55 reduced wet deposition. Finally, concentrations of organic carbon (OC) can be either
56 underestimated or overestimated in these two sites. The current model does not include
57 secondary organic aerosol (SOA) formation; inclusion will probably leads to higher simulated
58 OC concentrations.

59 For the evaluation of cloud properties, the simulated cloud fraction (CF) agrees fairly well with
60 observations over the ocean and in southern China, but significant underestimates occur in
61 northern China, especially in January. The domain average NMBs are -32% and -9% in January

62 and July, respectively. The liquid water path (LWP) is substantially underestimated across the
63 domain, with NMBs of -59% and -74% in January and July, respectively, which is a common
64 problem for many previous chemical transport model simulations^{2,5,12,13}. It is noted, however, the
65 LWP retrieved from MODIS may be biased by a factor of 2 due to uncertainties in cloud particle
66 size assumption¹⁴. The domain average cloud droplet number concentration (CDNC) is
67 moderately underpredicted by 29% and 38% in January and July, respectively, with the most
68 remarkable underestimates occurring on land in January. The discrepancies in cloud parameters
69 may be related to several factors including aerosol number concentrations, water vapor, aerosol
70 activation parameterization, cloud microphysics, and cumulus cloud schemes. For example,
71 Zhang et al.¹⁵ showed that the Abdul-Razzak and Ghan¹⁶ aerosol activation parameterization
72 used in this work tends to underpredict aerosol activation fraction and consequently underpredict
73 CDNC and LWP compared with the Fountoukis and Nenes¹⁷ parameterization. In addition, the
74 significant underestimates of CF and LWP in northern China may be explained in part by the
75 underestimates in water vapor mixing ratio. Finally, we note that large uncertainty in satellite-
76 retrieved LWPs^{14,18} and in CDNC estimation¹⁹ may also partly account for the model-
77 measurement discrepancy. Downward shortwave radiation at surface (SWD) and downward
78 longwave radiation at surface (LWD) simulated by WRF-Chem agree fairly well with the
79 CERES data in terms of both magnitude and spatial distributions, with NMBs of about 11% to
80 18% and -8% to -3% , respectively. The slight overestimates in SWD and underestimates in
81 LWD are likely induced by the underestimates in LWP and AOD.

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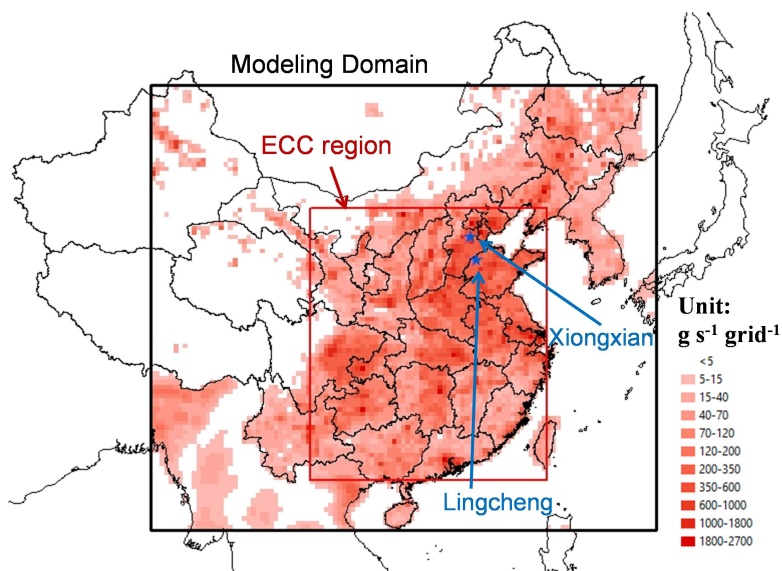
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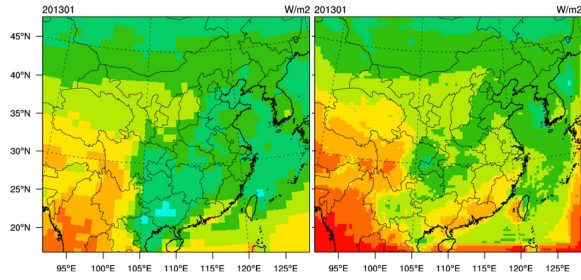
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149 **Supplementary Figure 1.** The WRF-Chem modeling domain at a horizontal spatial resolution
150 of 36 km (111 × 98 cells). The two blue stars indicate locations of the Xiongxian Site and the
151 Lingcheng Site, where observations of PM_{2.5} chemical components are available. The red
152 rectangle indicates the Eastern and Central China (ECC). This region has high population density
153 and high aerosol emissions and concentrations, and is thus the focus of this study. The colors
154 represent primary PM_{2.5} emission rates at 8:00 a.m. January 1st, 2013. This figure is produced
155 using DotSpatial, [version 1.7], (<http://dotspatial.codeplex.com/>) and Microsoft PowerPoint 2013
156 (<https://www.microsoft.com/>).

Observation in January

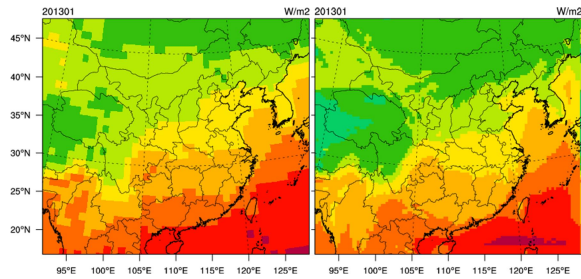
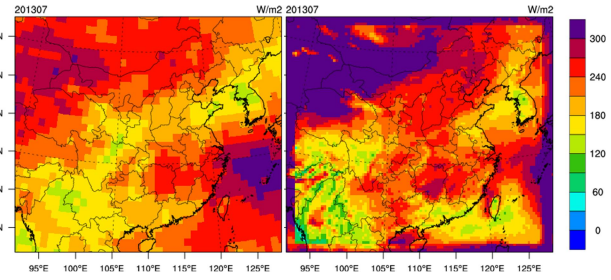
Simulation in January



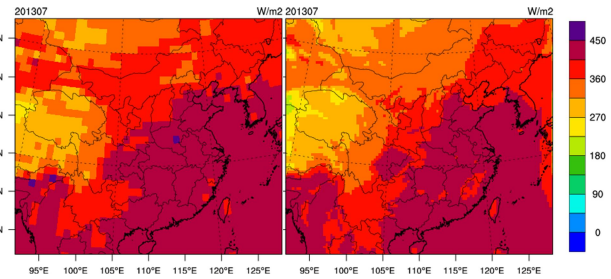
Observation in July

Simulation in July

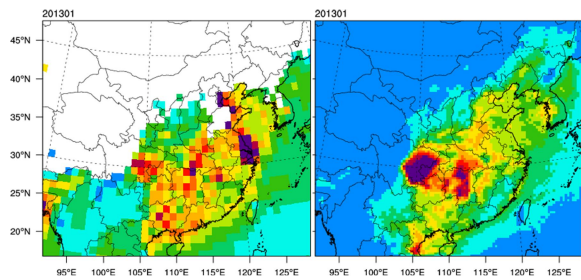
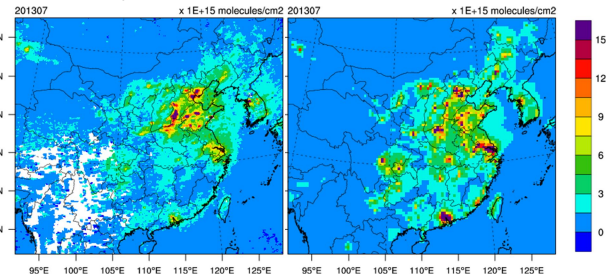
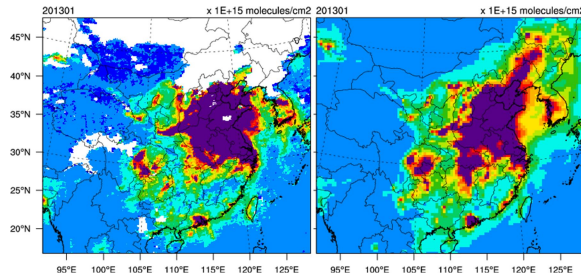
SWD



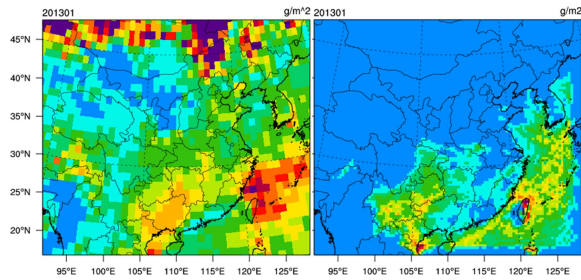
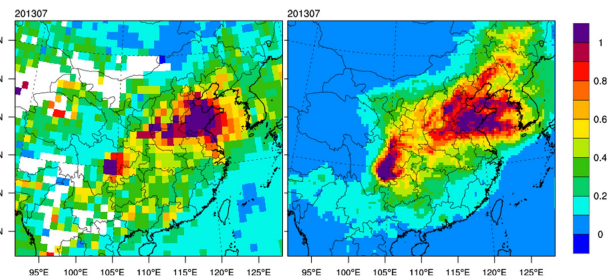
LWD



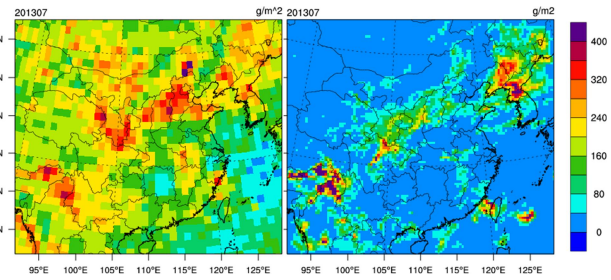
NO₂ vertical column density



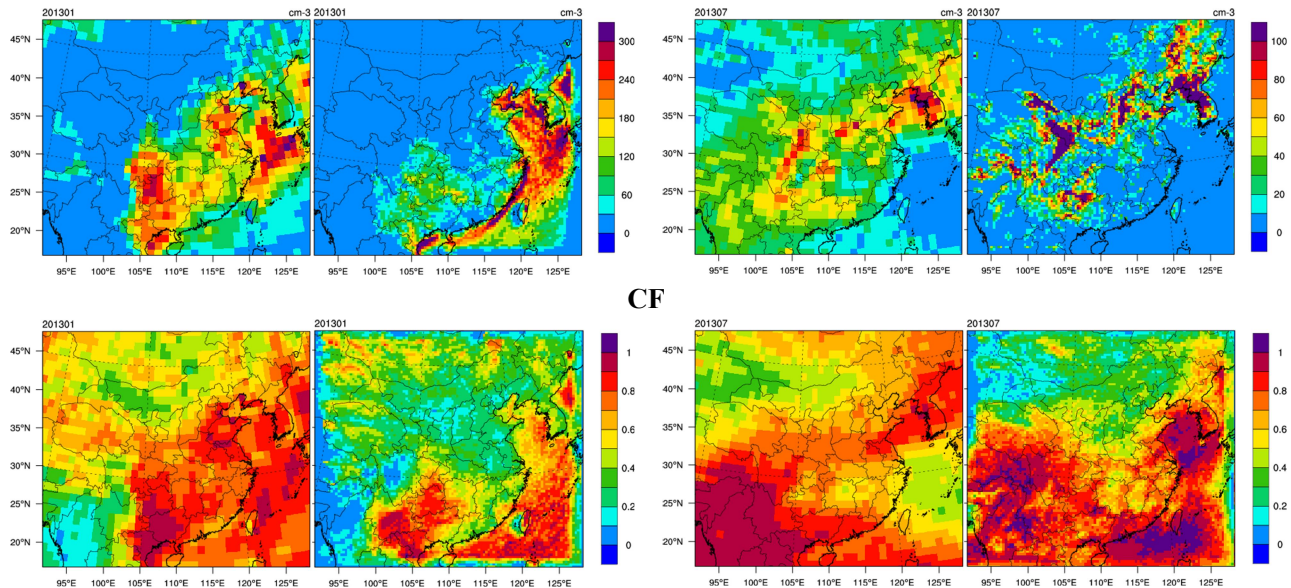
AOD



LWP



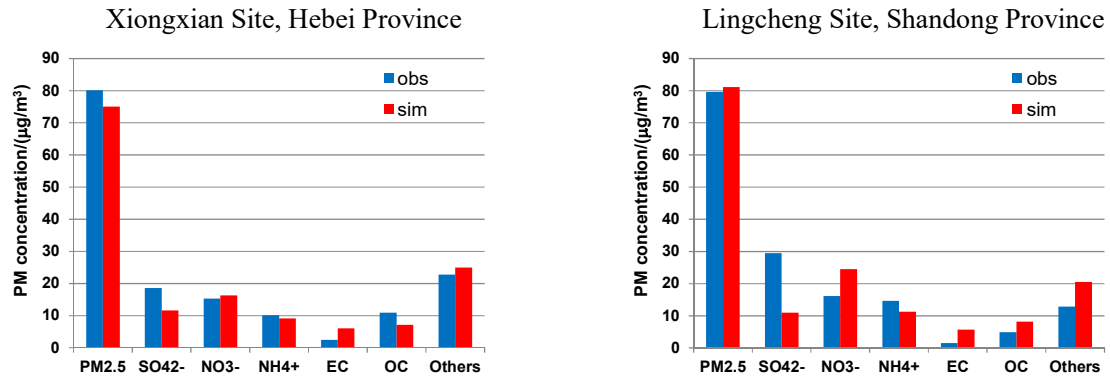
CDNC



CF

157 **Supplementary Figure 2.** Comparison of simulated SWD, LWD, NO₂ column, AOD, and cloud
 158 properties with satellite observations. This figure is produced using the NCAR Command
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162 **Supplementary Figure 3.** Comparison of simulated concentrations of PM_{2.5} and its chemical
 163 components with observations at the Xiongqian Site and the Lingcheng Site. This figure shows
 164 the average concentrations during July 22nd - July 31st.

165 **Supplementary Table 1.** Observational datasets used in model evaluation.

Type	Database	Variables	Sites/resolution	Frequency
Surface meteorology	NCDC	wind speed at 10 m (WS10), temperature at 2 m (T2), and water vapor mixing ratio at 2 m (Q2)	380 sites	Hourly or every 3 hour
	GPCC	precipitation	0.5°×0.5°	Monthly
Surface air quality	MEP	PM ₁₀ , PM _{2.5} , SO ₂ , NO ₂ , and O ₃	496 sites in 74 large cities in China	Hourly
	PM _{2.5} chemical components	PM _{2.5} , SO ₄ ²⁻ , NO ₃ ⁻ , NH ₄ ⁺ , BC, and OC	2 sites: Xiongxian Site, Lingcheng Site	Daily
Satellite	CERES	Downward shortwave radiation at surface (SWD), downward longwave radiation at surface (LWD)	1°×1°	Monthly
	OMI	NO ₂ vertical column density	0.125°×0.125°	Monthly
	MODIS/TERRA	aerosol optical depth (AOD), liquid water path (LWP), cloud fraction (CF), and cloud droplet number concentration (CDNC, derived from MODIS data)	1°×1°	Monthly

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